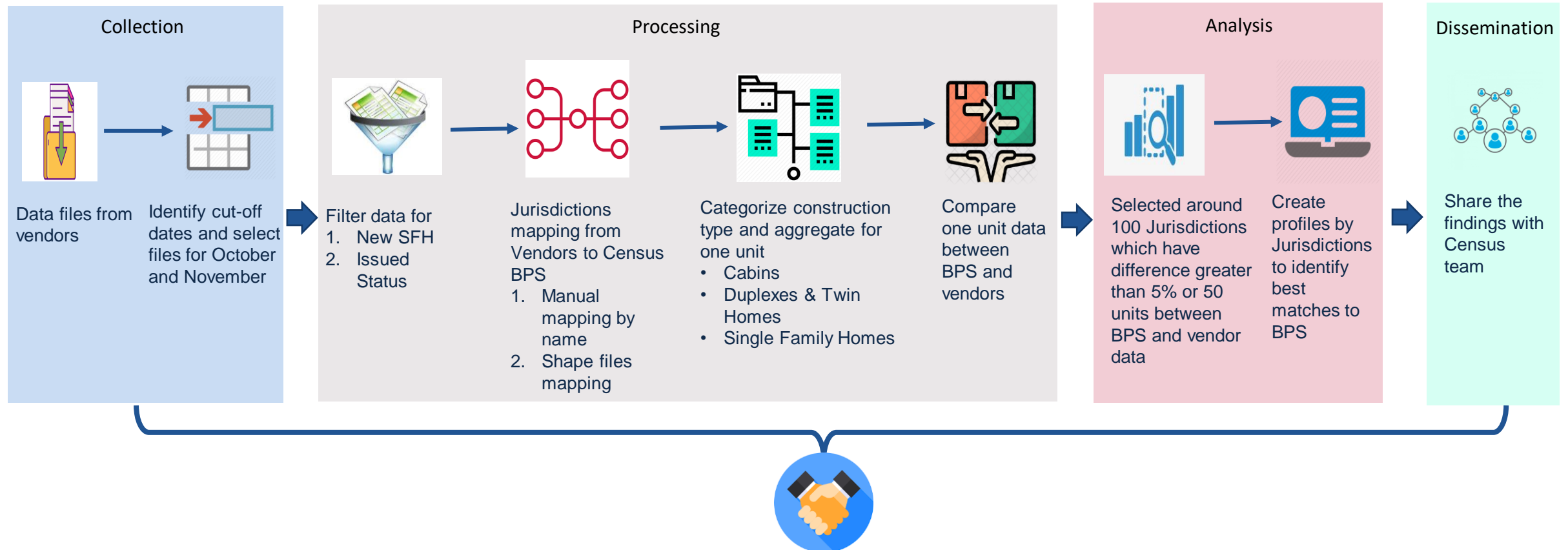


Construction Re-engineering Presentation to CSAC

March 19, 2021

Build BPS with 3rd Party Data Sources



Continuous collaboration with the Census Survey Stats & Math Stats Team

Data Collection

Construction
Monitor

Weekly data delivery on
every Monday through FTP
starting Oct 13th



Select cut-off dates and
Pick files from week 41-
47 for October and week
45-51 for November

CM	Jurisdictions covered	Records count
Oct	1643	164 k
Nov	1485	128 k

CoreLogic

Weekly data delivery on
every Thursday through
FTP starting Nov 16th



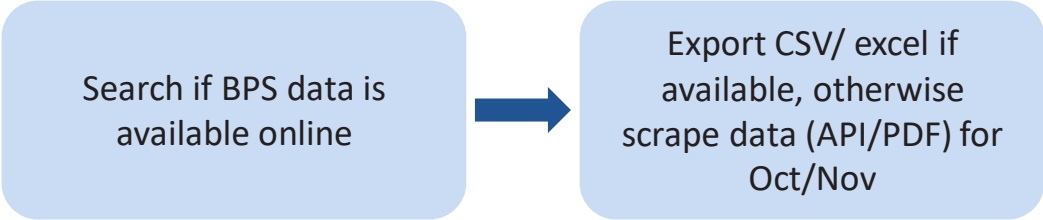
Select cut-off dates and
Pick files for October and
November

CL	Jurisdictions covered	Records count
Oct	458	137 k
Nov	201	88.5 k

The jurisdictions covered by 3rd party data sources account for about 70% of single-family units authorized in the U.S.

Data Collection

Website



Permit/Application Status Search

Jurisdiction:* Snohomish County * - Indicates a required field

Search by:

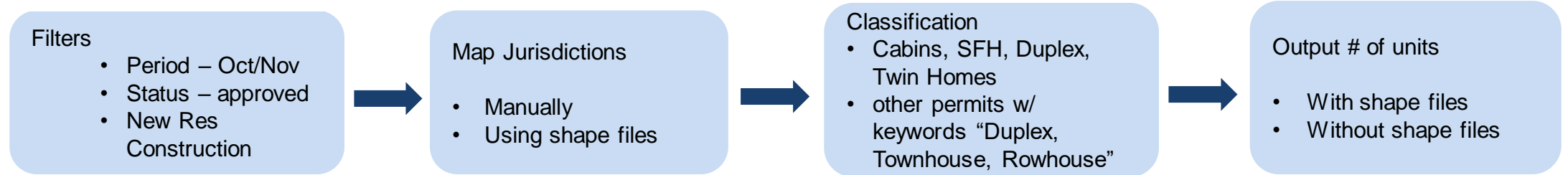
Permit #	Project Info	Location	People
Project Name/Description (partial match): <input type="text" value="Enter the name of the project"/>			
Permit Type: ⓘ	RK - Residential Building ✕		
Permit Status: ⓘ	-- Select Statuses --		
Date Type:	Issued ✕		
From:	10/01/2020		
To:	10/31/2020		
ⓘ Jurisdiction is a required field Search Cancel			

☒ Export to Excel

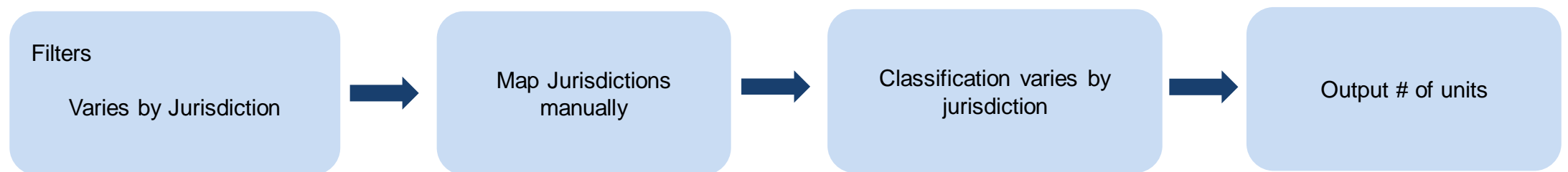
	Permit #	Description	Address	Type	Status	Applied Date	
▶	17107952RK	Expired without final inspectio...	15927 73RD AVE NE	RK	Closed	05/19/2017	
▶	18127354RK	Application Expiration date ext...	12904 227TH AVE SE	RK	Issued	11/18/2018	
▶	19115079RK	Application Expiration date ext...	13606 S MACHIAS RD	RK	Issued	10/10/2019	
▶	19117740RK	Application Expiration date ext...	3124 ROBE MENZEL RD	RK	Issued	11/25/2019	
▶	19117989RK	Application Expiration date ext...	1303 139TH PL SW	RK	Issued	12/06/2019	

Data Processing

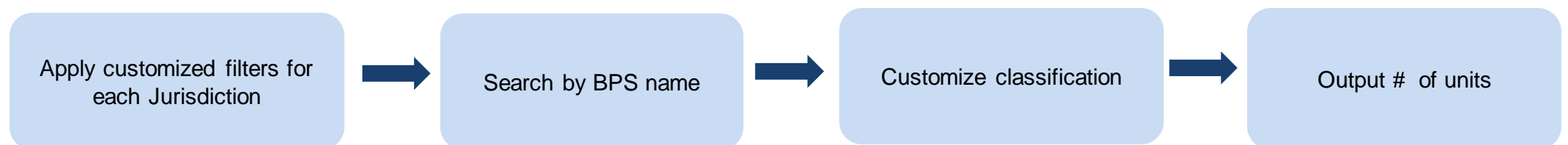
Construction Monitor



CoreLogic



Website



Next Steps (...over next 3-4 months)

- Expansion of jurisdictions
- Jurisdiction mapping based on Census geolocations
- Classification of construction types library
- Add Multi-family homes to BPS work stream

Applying and AI and Satellite Imagery for Construction Classification

Objective and Success Criteria

Goal

- Demonstrate potential value of the identification and categorization of construction sites using satellite images.



Definition of construction starts:

- *Excavation and delimitation of building site*
- *Laying the concrete foundation of the building*

Success Measures

- **Cost:** Demonstrate that construction starts and completions can be accurately estimated using satellite imagery at a cost equal to or less than existing methods
- **Timeliness:** Demonstrate through the POC that monthly building starts can be estimated using geospatial images and ML/AI as per existing program timelines

Initial Project Phases

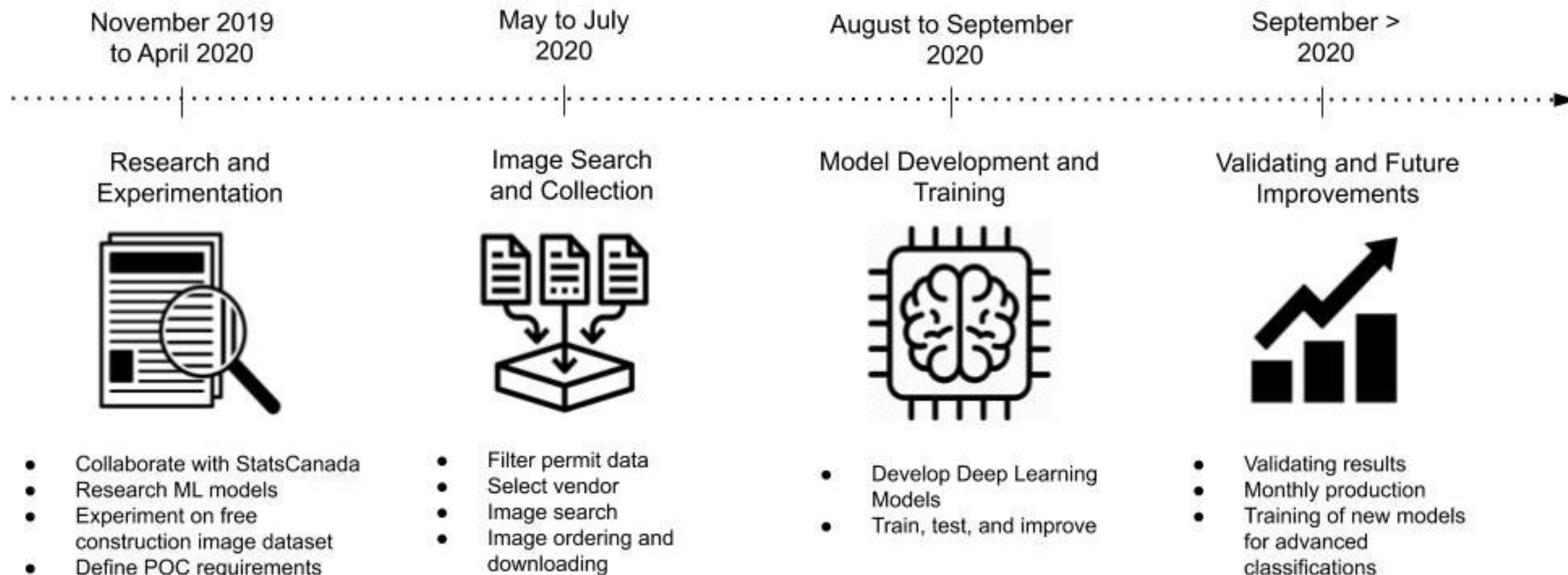
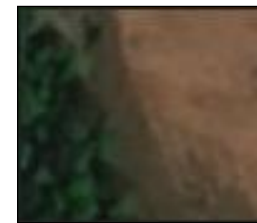
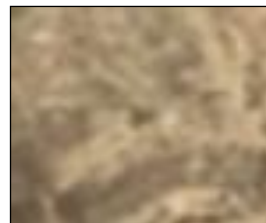


Image Categorization

Pre-constructions

90 days before permit authorization date or earlier



This image category displays the construction stage before any construction has happened. Images display the ground untouched and no major delimitations or excavations.

Construction Starts

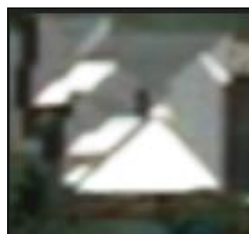
Between 30 days after permit authorization date and 125 days after it.



This image category displays the construction stage immediately after construction has started. Our defined “start” definition is: “visible excavation or foundation”.

Construction Completions

270 days after permit authorization date.



This image category displays the completed roof covering the area where there was a foundation or excavation previously.

CNNs Development Mode

Model A (FastAI)

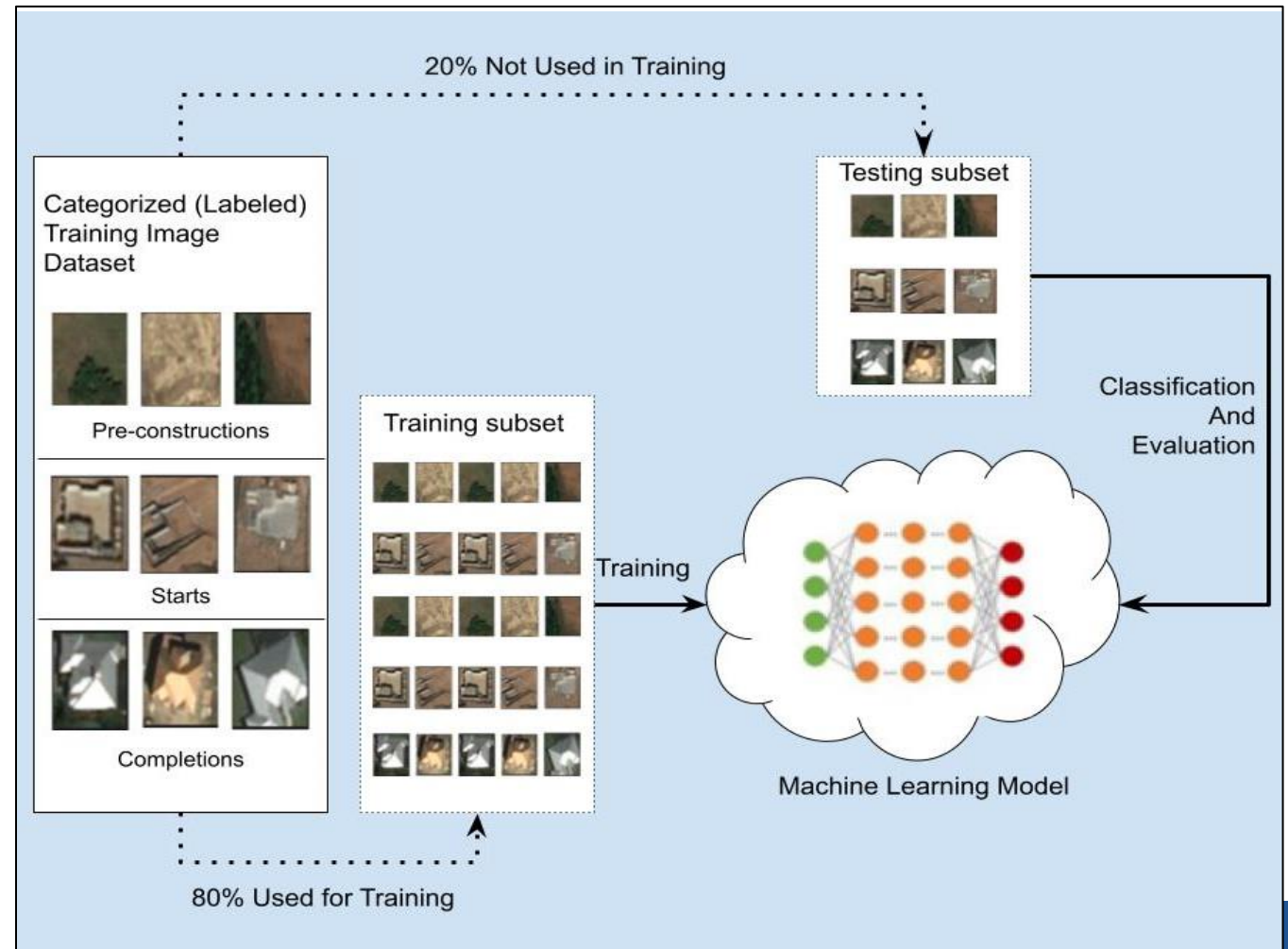
We utilized Fast.ai libraries and ResNET, a pre-trained Deep Learning model for image classification most applied to analyzing visual imagery. StatsCanada has shown promising results using similar methodology and tools.

- Rapid Prototyping
- Best utilization and optimization for GPUs

Model B (Keras)

Keras is a deep learning API written in Python, running on top of the machine learning platform TensorFlow. It was developed with a focus on enabling fast experimentation. Being able to go from idea to result as fast as possible is key to doing good research.

- Large-scale deployments
- Cross-platform



Training and Testing

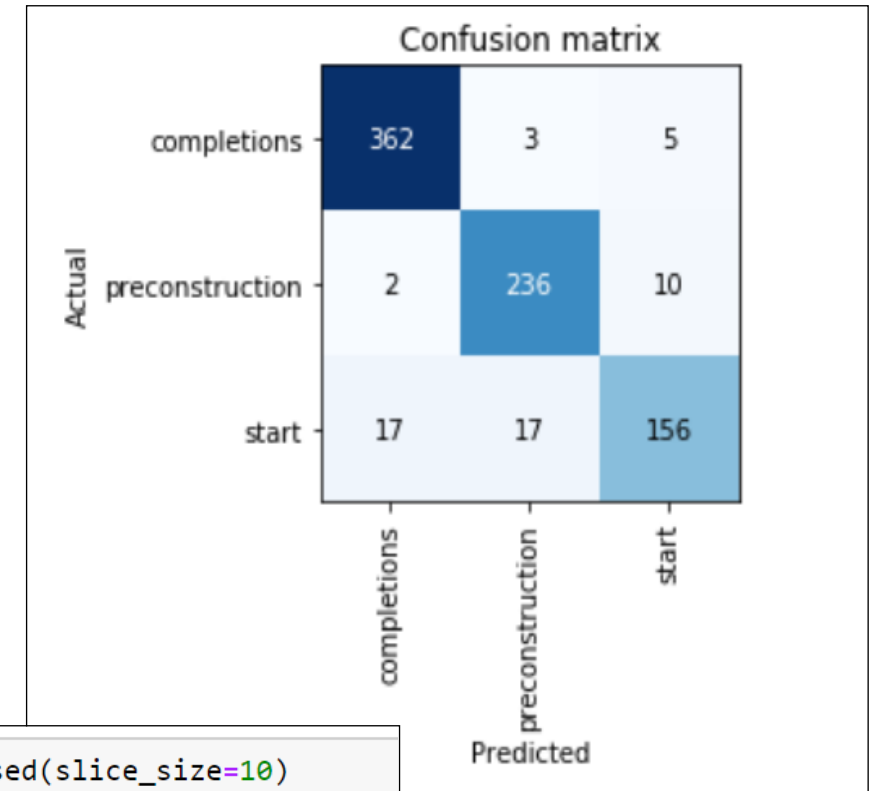
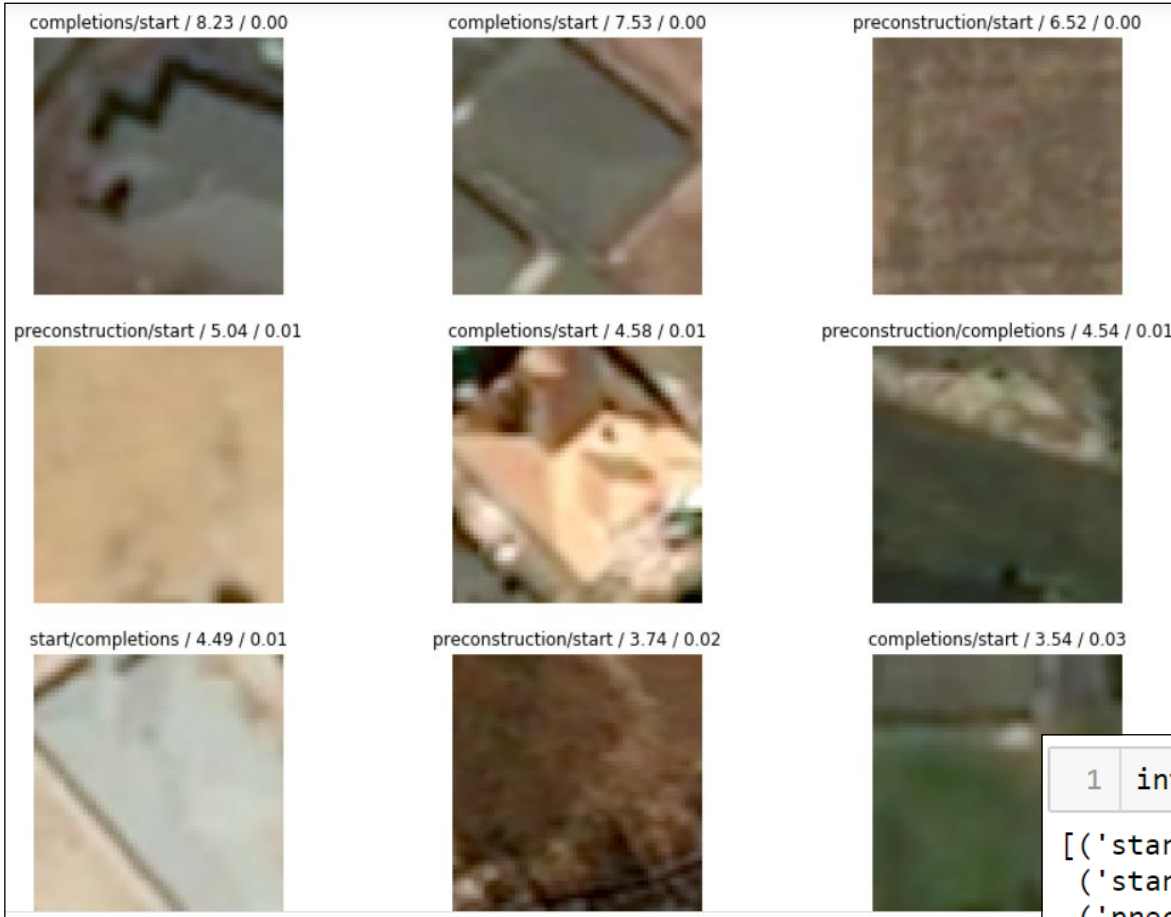
The **FastAI model** was trained to classify between 3 construction stages image categories (precons, starts, and completions). It has been trained with 3236 images in the training subset and 808 images in the relative validation subset. Initial values of the fitting, showed 8% error rate (or 92% accuracy score). **After optimizing the FastAI model we achieved 94% accuracy score.**

epoch	train_loss	valid_loss	error_rate	time
0	0.114260	0.310744	0.084158	00:26
1	0.172425	0.280688	0.081683	00:27
2	0.158284	0.250213	0.077970	00:29
3	0.123952	0.245381	0.082921	00:29

For **Keras model**, the team used 3,262 images as the training set and 814 images as the validation set. The images are labeled as “Complete” and “Incomplete”. **The model has achieved an accuracy score as high as 93%** during the training process, with an average accuracy score around 90%. As we further train/ tune the model with more images from our vendors, we will be able to continue to improve our model performance.

epoch	train_loss	valid_loss	error_rate	time
0	0.100892	0.211188	0.077970	00:24
1	0.099556	0.206739	0.075495	00:25
2	0.096449	0.208337	0.077970	00:24
3	0.097957	0.205757	0.066832	00:24

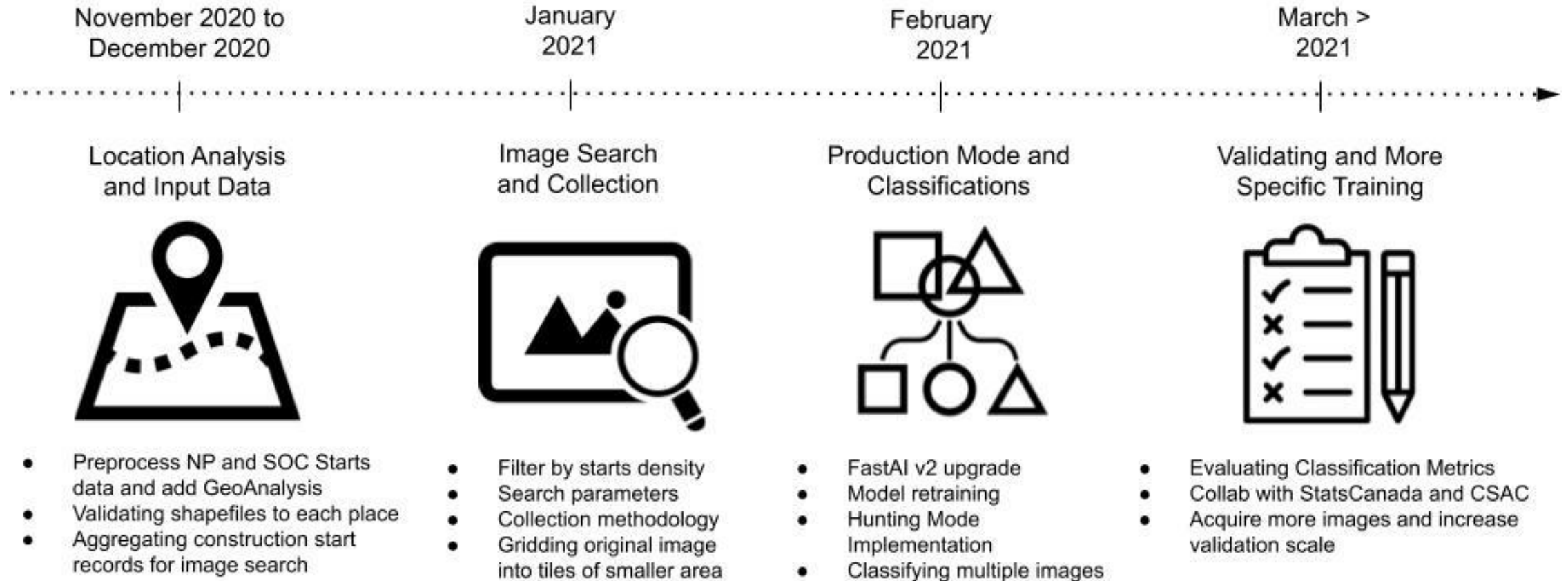
Validating Models



```
1 interp.most_confused(slice_size=10)
```

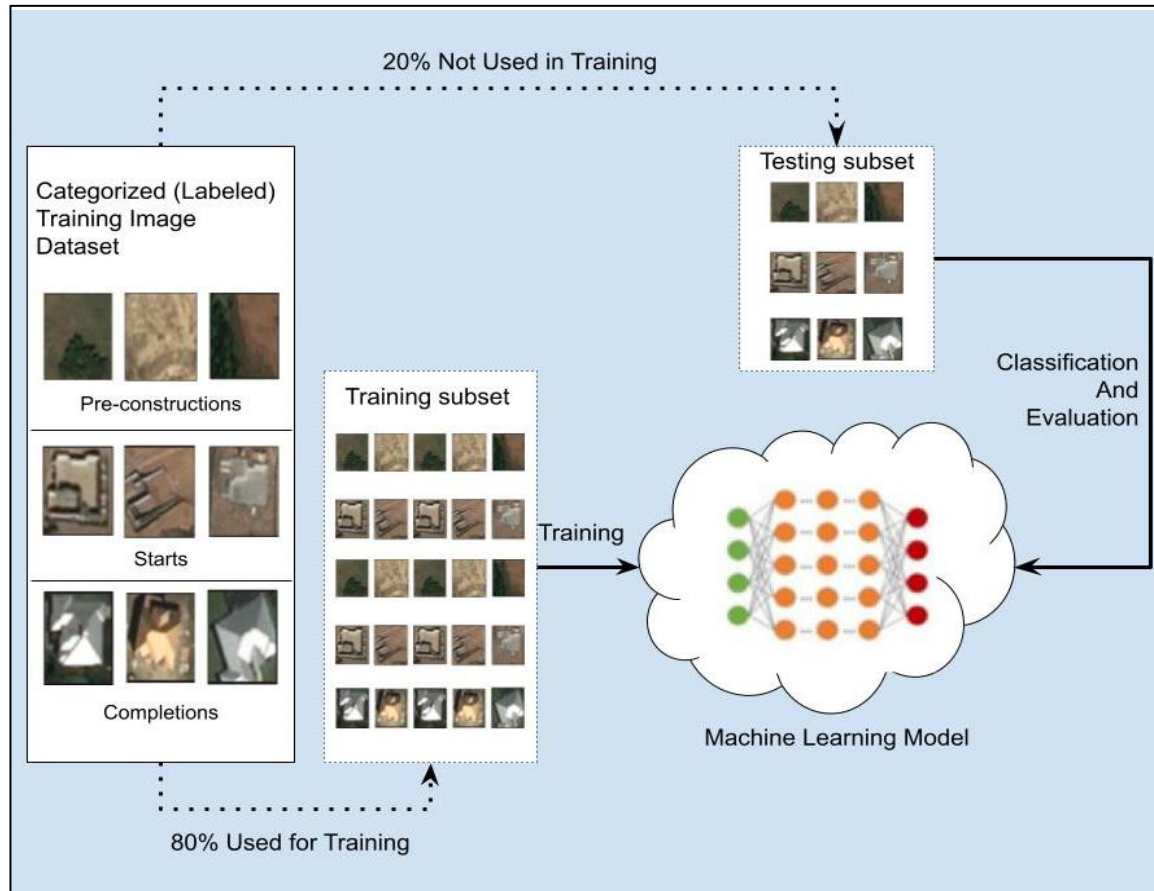
```
[('start', 'completions', 17),  
( 'start', 'preconstruction', 17),  
( 'preconstruction', 'start', 10),  
( 'completions', 'start', 5),  
( 'completions', 'preconstruction', 3),  
( 'preconstruction', 'completions', 2)]
```


Current Project Phases

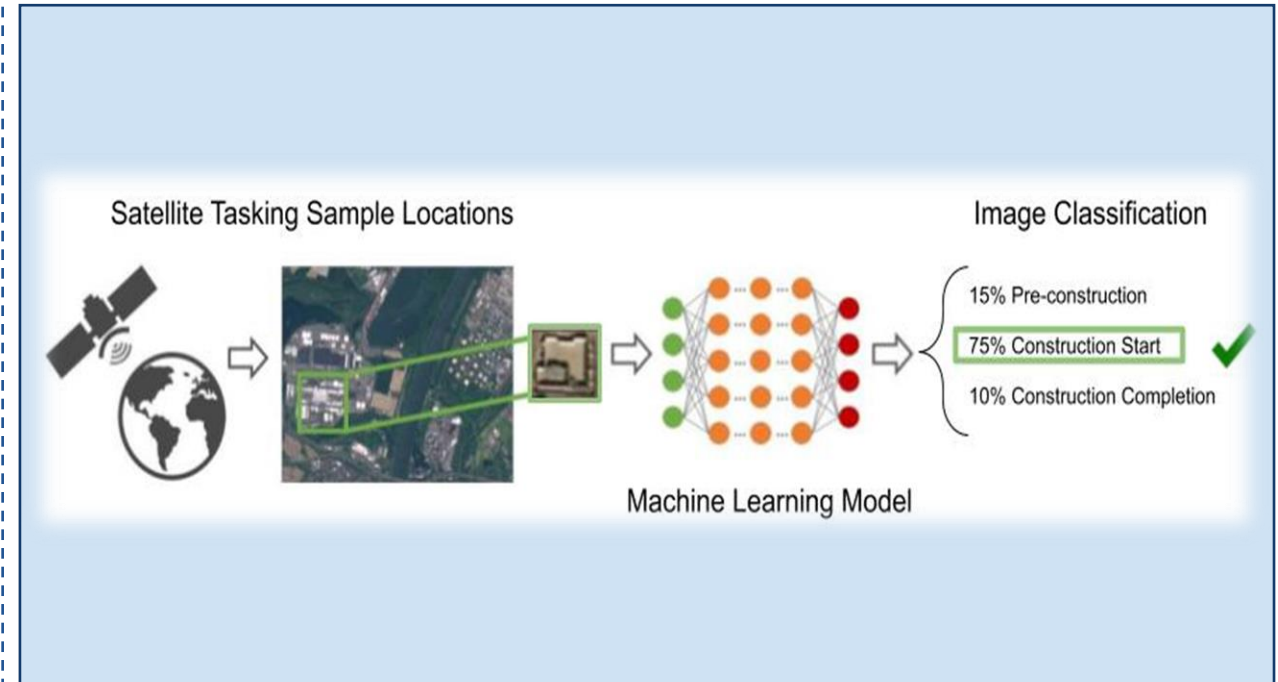


From Dev to Production

Training and Testing Model



Model used in Hunting Mode



Mapping Construction Starts Data

- Pre-process Construction Activity Data
- Shapefile Matching and Spatial analytics

- Shapefiles matches
- Aggregating records by location, data, and construction starts count

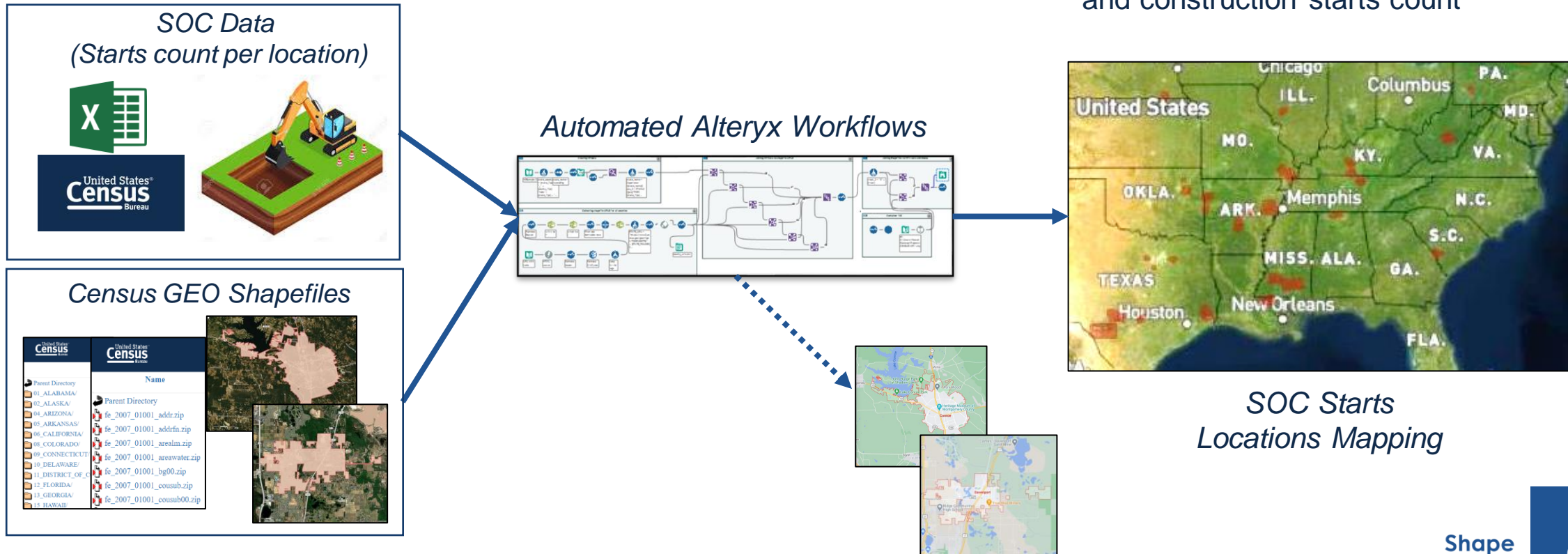
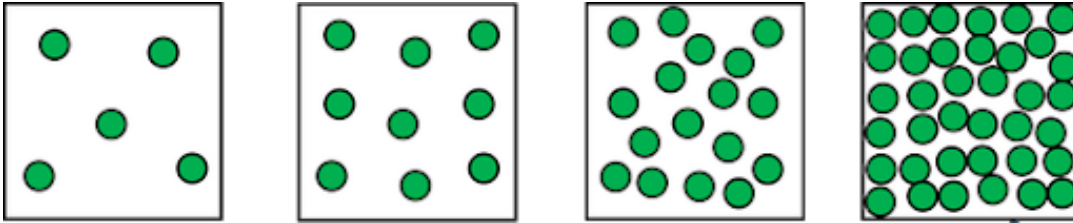
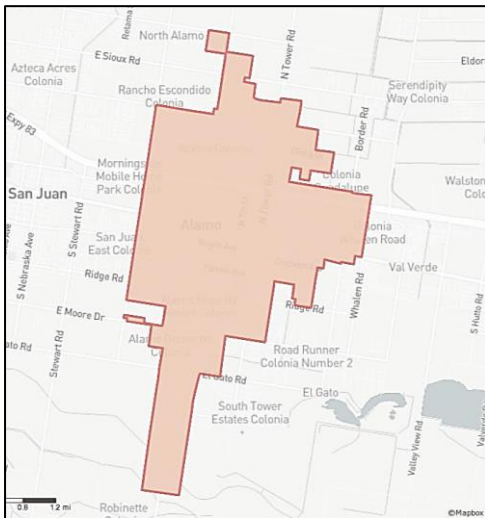


Image Data Collection

- Filters by place construction density: (starts count / area in sq km)



- Bounding Boxes



- Image search parameters

< 15% Cloud coverage



Image fully covers shapefile

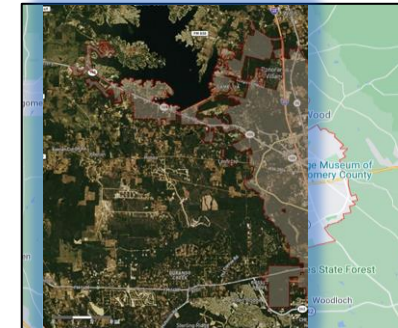
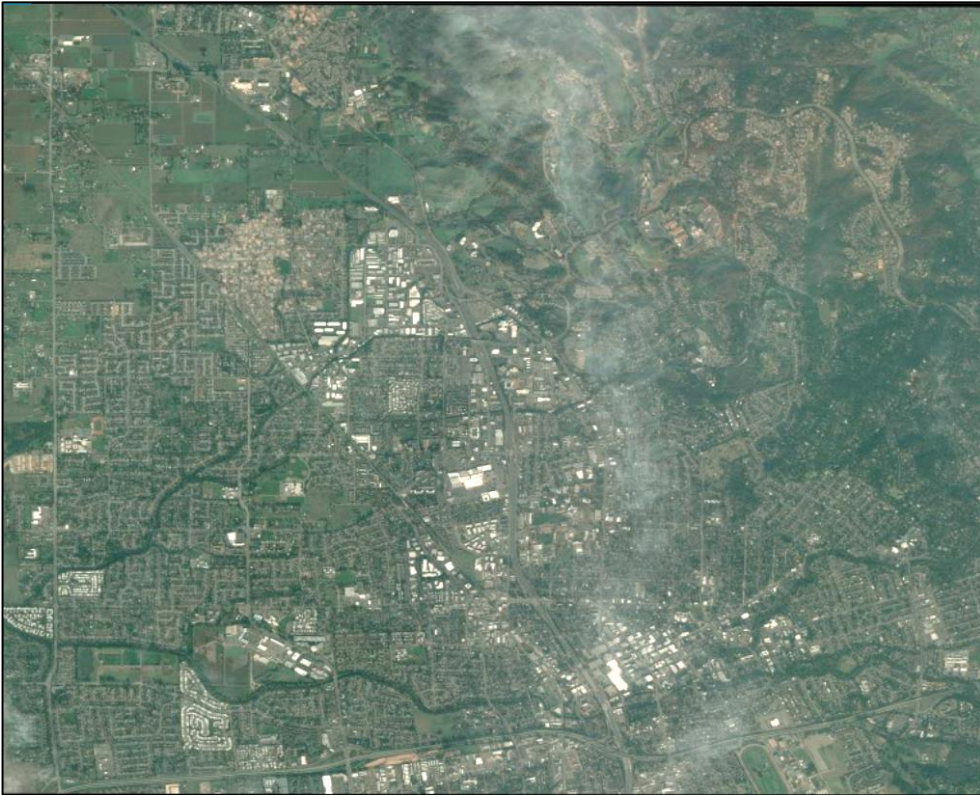


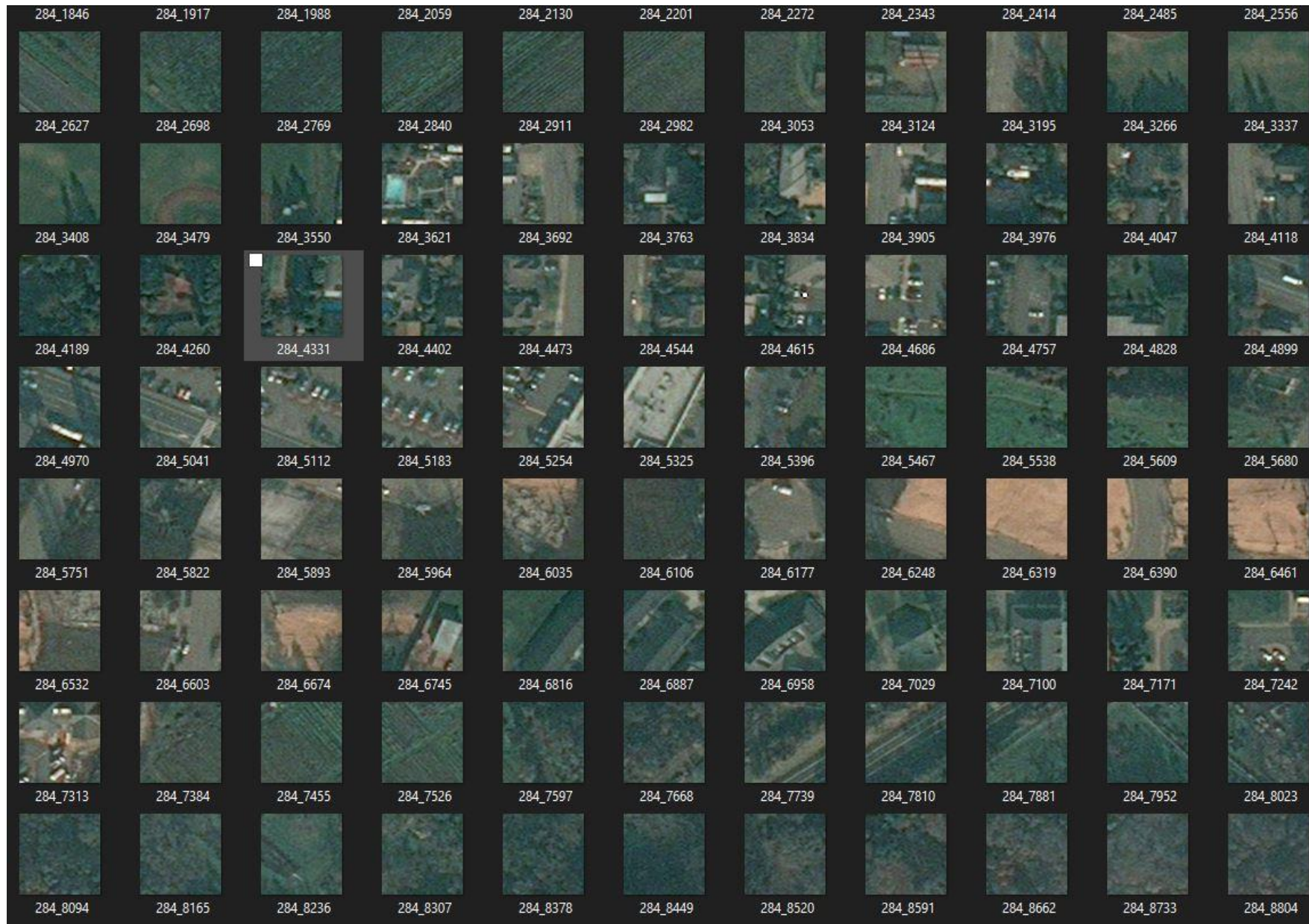
Image Collection & Grid Methodology

- Image search selected sample



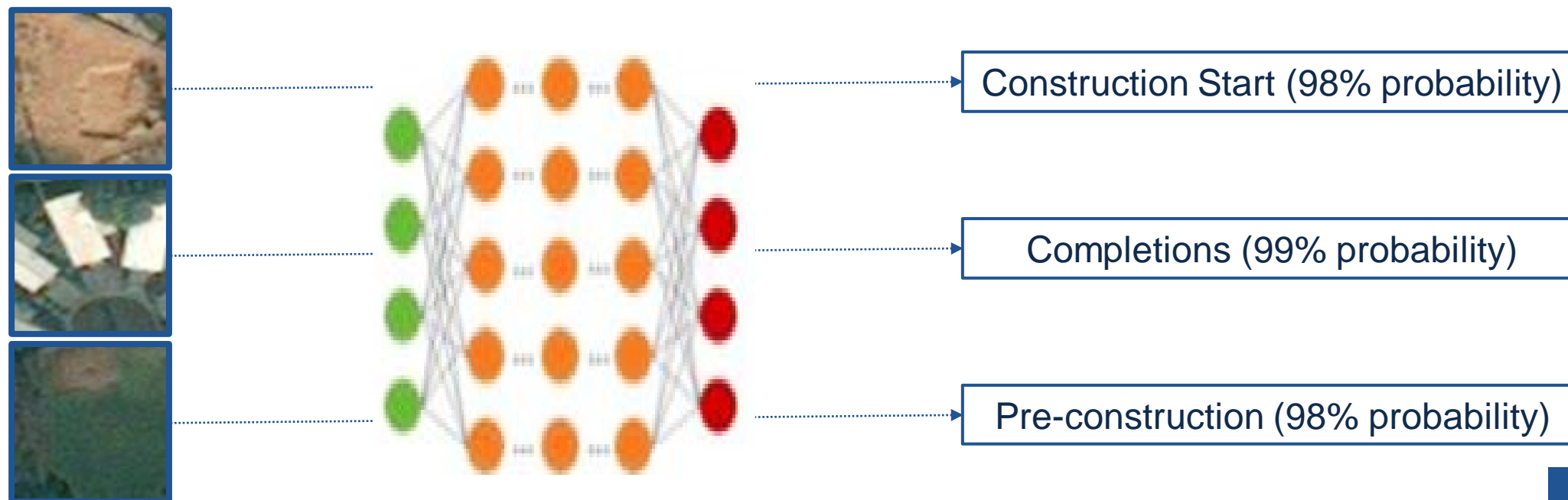
- Gridding into 0.001 sq km tiles





Production Mode

- FastAI v2 on a new **dedicated** Reveal Linux server
- Model re-trained on server (GPU enabled)
- Creating production mode environment: from grids to answer

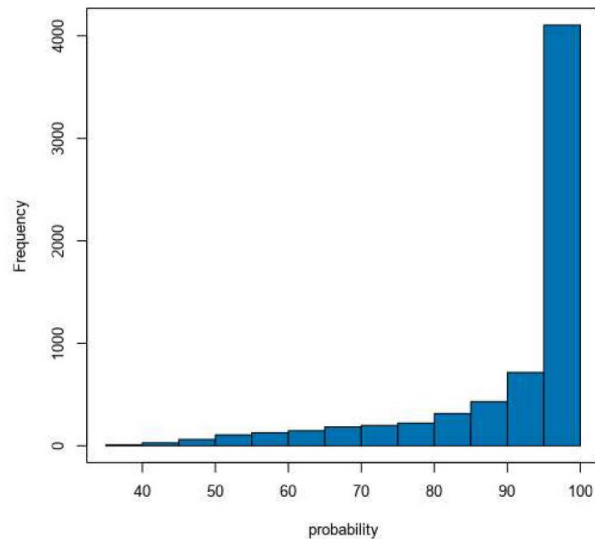


Construction Classifications

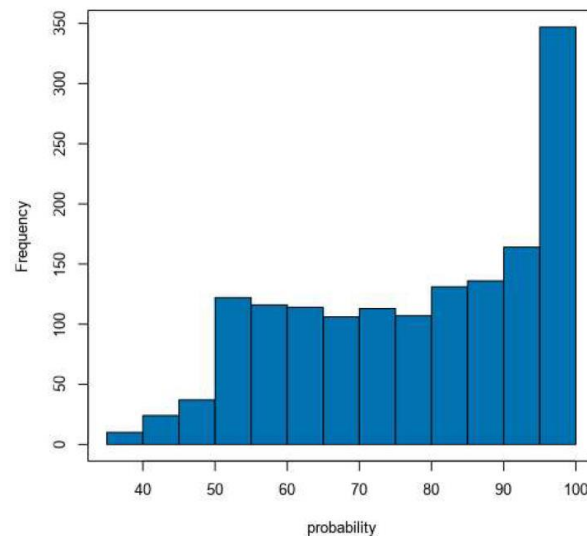
- Classification Process and Metrics
 - 9409 tiles (images) classified
 - 13 minutes classification time

Category	Avg_probability	Max_probability	Min_probability
Completions	82.526856	99.9999	36.6909
Preconstructions	91.195509	99.9998	35.5002
Start	77.87969	99.9999	35.6627

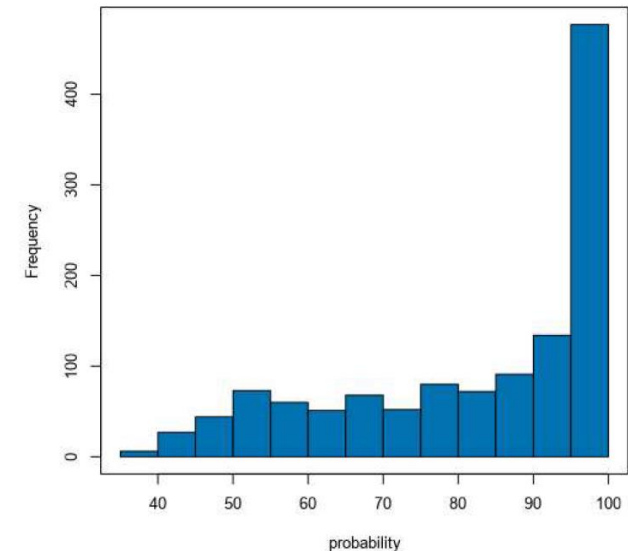
Preconstructions



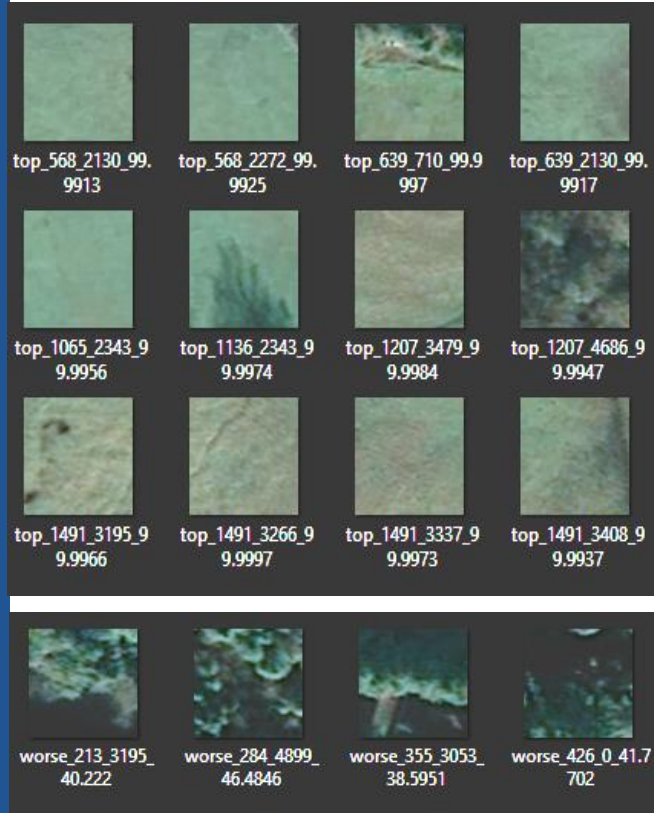
Starts



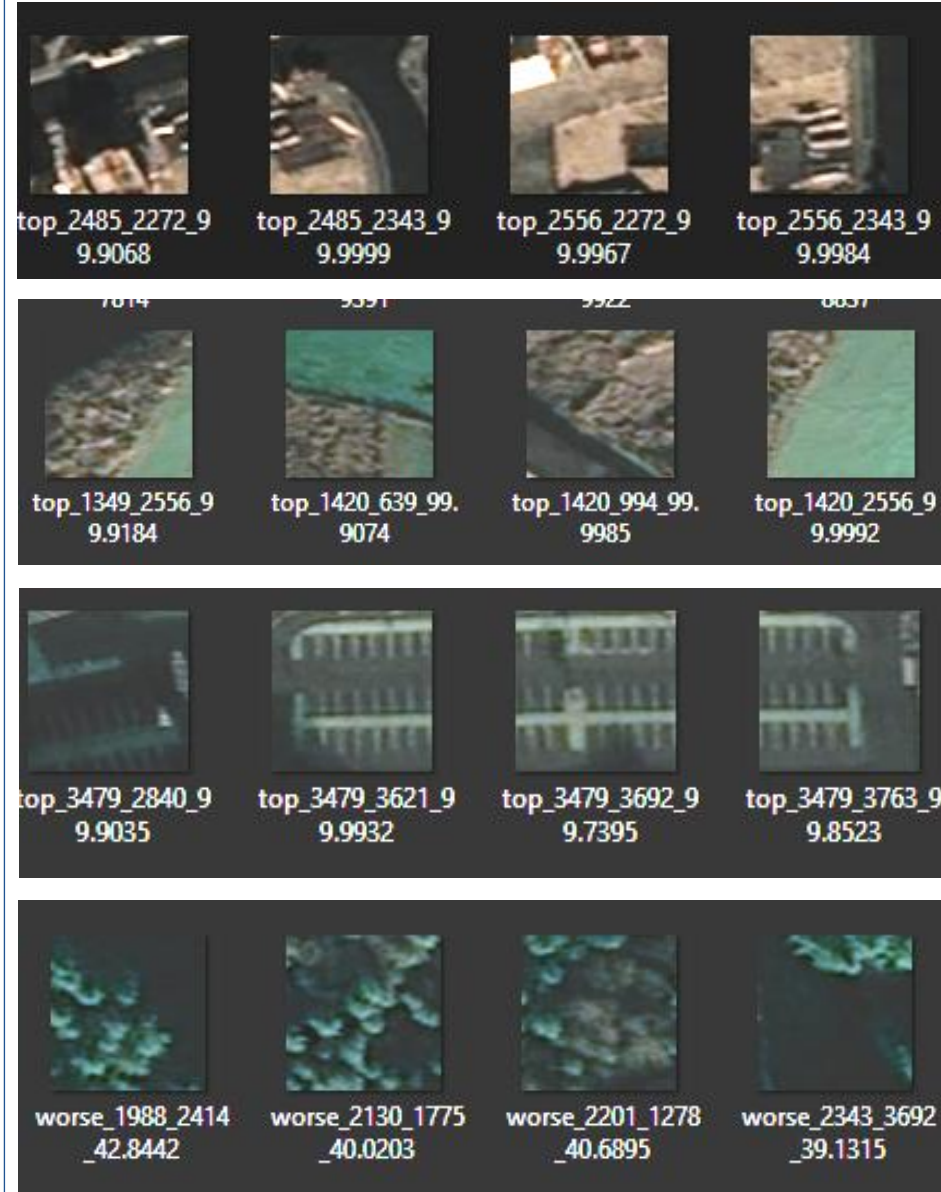
Completions



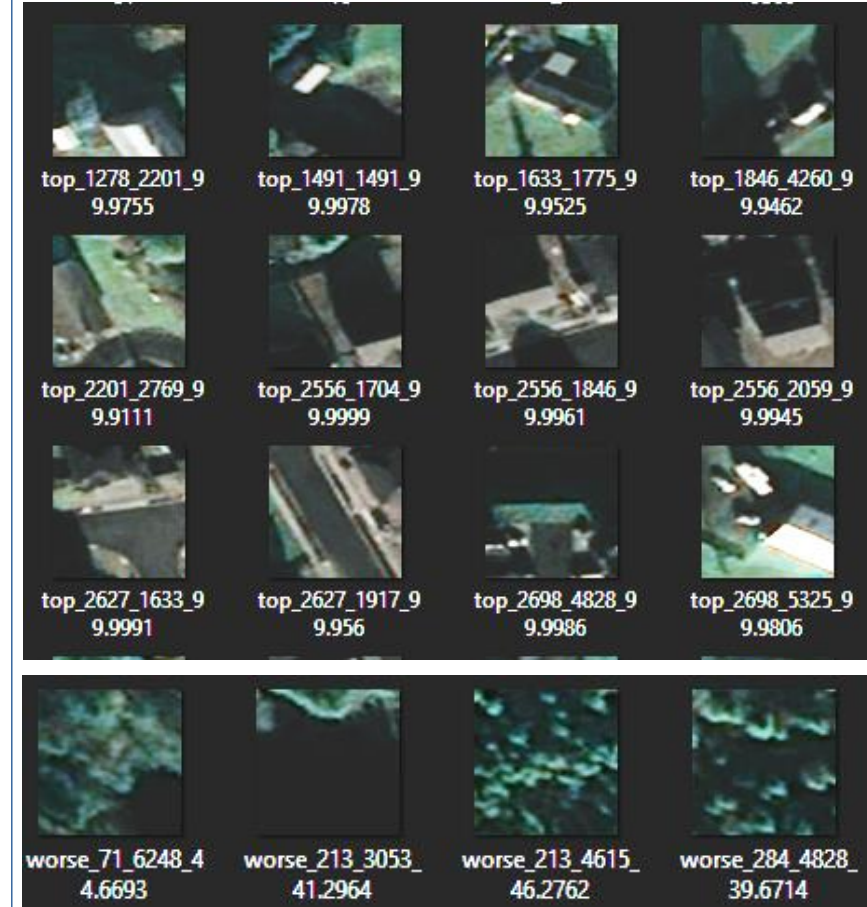
Precons



Starts

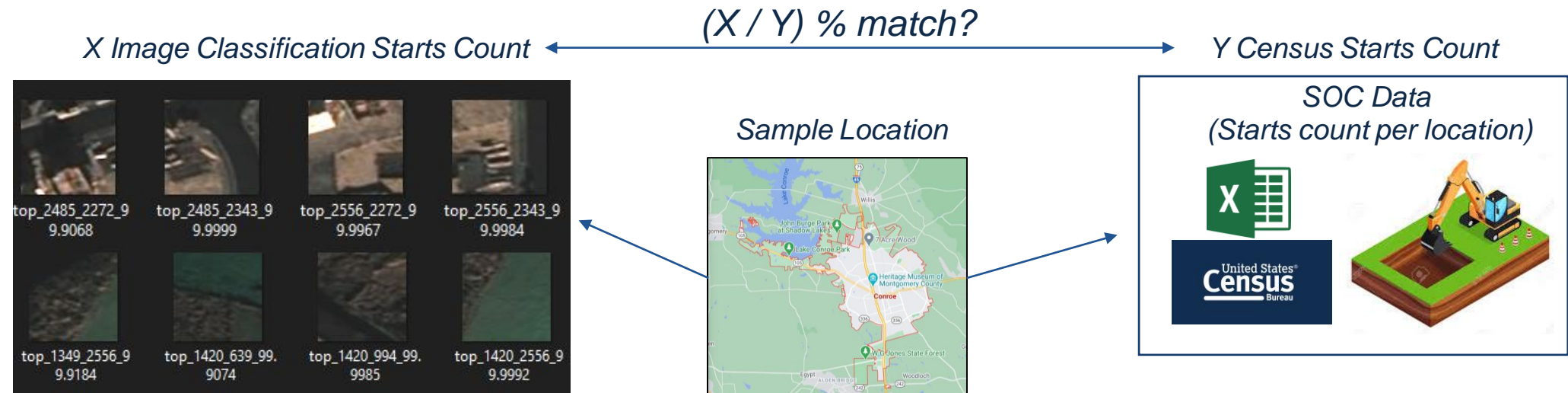


Completions



Hunting Mode Validation

- Taking results from classifications
- Aggregating based on locations (geoid)
- Matching aggregation against Census Data

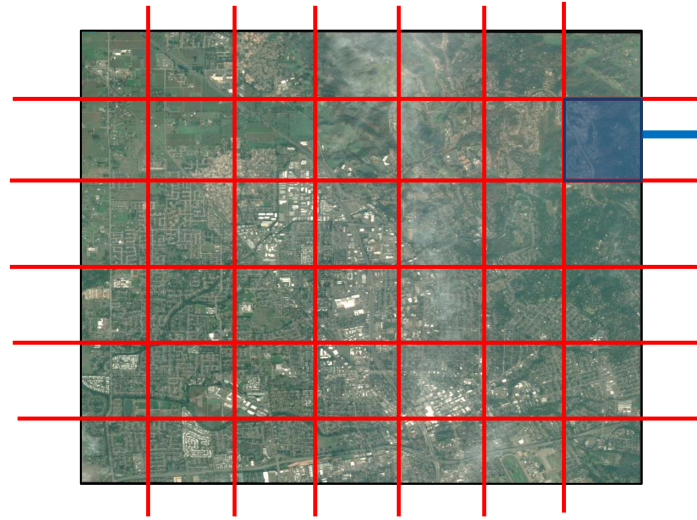


Revised Grid Methodology

- Collect image for selected location



- Grid to bigger tile area of 0.1sqkm

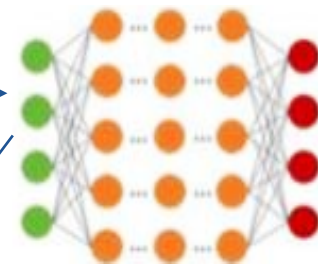


- Automatic Construction Boundary Detection

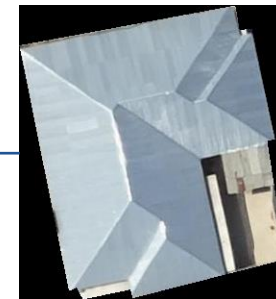


Start

Classification Model

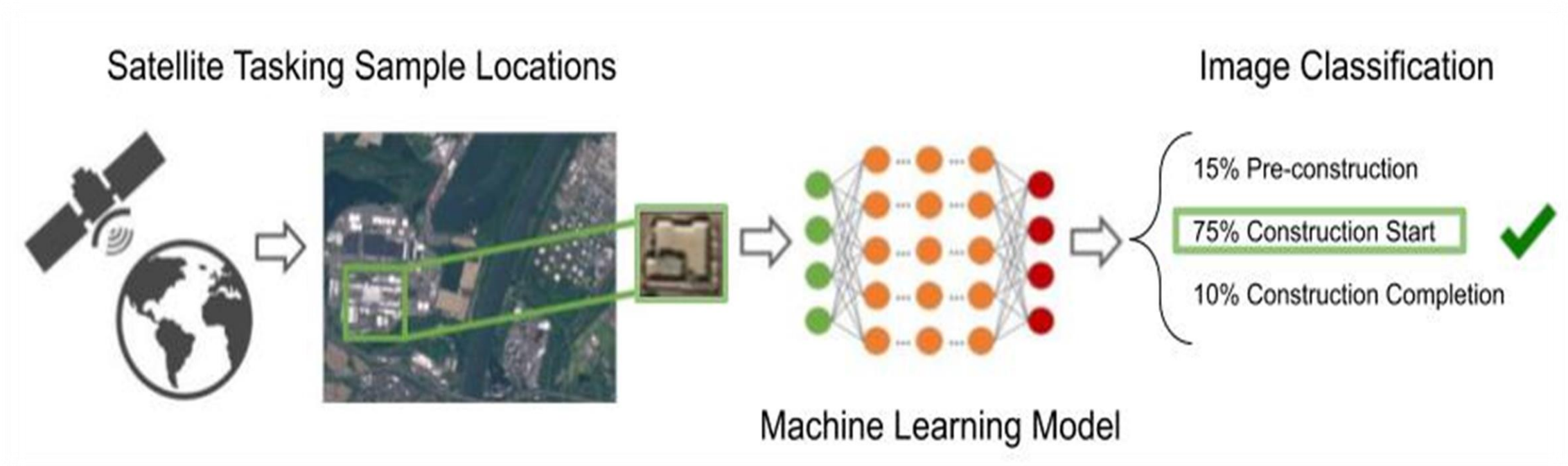


Completion



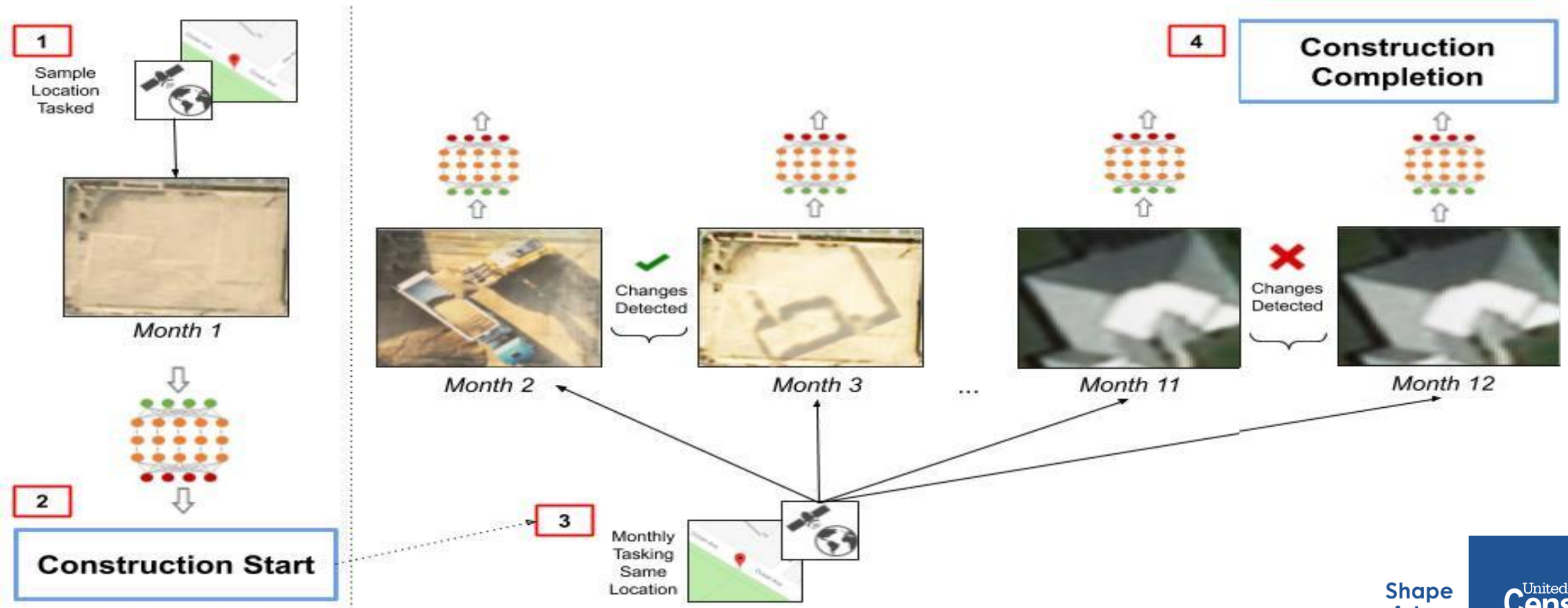
Hunting Mode

After refining our training image dataset, we decided to use two modalities for operations of the satellites. In “Hunting Mode”, images are collected monthly and fed to the CNN models to detect if there are any construction starts.



Tracking Mode

Once a construction start has been detected in an image, we will track that location on a monthly basis for changes. We will apply change detection techniques to track construction progress and tag the construction activity as complete based on the detection of pre-established features such as finished roof, driveway, landscaping, absence of construction vehicles, etc.



Change Detection Implementation

Change Detection algorithm in between construction stages is already implemented and tested, ready for tasked images.

Starts



Completions



Next Steps

- GUI interface for human validation and model feedback
- Classify more SOC Starts locations and test Non-Permit locations
- Automatic construction boundary detection
- Training new non-construction categories to filter out false positive starts:
 - Streets, parking lots, trees, pools
- Specialize current construction categories:
 - urban starts, rural starts
- Integrate Hunting and Tracking mode
- Train model on residential multi-unit, and on commercial buildings
- Train model to infer construction \$ value